Computing Algebraic Immunity by Reconfigurable Computer

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Abstract

Algebraic immunity (AI) is a property of a Boolean function f that measures its susceptibility to an algebraic attack. If f has a *low* algebraic immunity and f is used in an encryption protocol, then there are ways to successfully cryptanalyze the system. As a result, it is important to have an efficient means to compute the algebraic immunity of Boolean functions. Unfortunately, algebraic immunity is one of the most complex cryptographic properties to compute. For example, it is significantly more difficult to compute than nonlinearity [2]. Here, we show the advantage of a reconfigurable computer in computing a function's algebraic immunity. For example, we show that a reconfigurable computer is 4.9 times faster than a conventional computer in this computation for 5-variable functions. Indeed, we compute the distribution of functions to algebraic immunity for all 5-variable functions, a computation that has not been previously accomplished. Interestingly, the problem we address is to design a logic circuit that computes a characteristic of a logic function.

1 Introduction

Any stream or block cipher can be described by a system of equations expressing the ciphertext as a function of the plaintext and the key bits. An algebraic attack is simply an attempt to solve this system of equations for the plaintext. If the system happens to be overdefined, then the attacker can use linearization techniques to extract a solution. However, in general, this approach is difficult, and not effective, unless the equations happen to be of low degree. That is (somewhat) ensured if, for instance, the nonlinear Boolean function combiner in an LFSR-based generator (a widely used encryption technique) has low degree or the combiner has a low algebraic immunity (defined below) [4, 5].

Let $\mathbb{F}_2$ be the two-element field and $\mathbb{V}_n = \mathbb{F}_2^n$ be the vector space of dimension n over $\mathbb{F}_2$, consisting of n-bit tuples, with the usual vector operations. Let an LFSR be filtered by a Boolean function $f(x_1, \ldots, x_n)$ of degree d, where another function $L : \mathbb{F}_2^n \to \mathbb{F}_2^n$ defines the LFSR. Suppose the keystream $z_0, z_1, z_2, \ldots$ is computed from some initial secret state (the "key") given by n bits $a_0, a_1, \ldots, a_{n-1}$ in the following way. Let $\mathbf{a} = (a_0, a_1, \ldots, a_{n-1})$ be the initial state vector and define the keystream bits by

$$
\begin{aligned}
z_0 &= f(\mathbf{a}) \\
z_1 &= f(L(\mathbf{a})) \\
z_2 &= f(L^2(\mathbf{a})) \\
\cdots \quad &\cdots \quad \cdots\cdots\cdots \\
z_t &= f(L^t(\mathbf{a})).
\end{aligned}
$$

The problem of extracting the plaintext message in this context is equivalent to the problem of finding the initial key $\mathbf{a}$, knowing L and f, and intercepting z_i. Since $\deg(f) = d$, every term on the right hand side of any of the above equations is one of the monomials made up of a product of some subset of d or fewer of the unknowns $\mathbf{a}$. There are $M = \sum_{i=1}^{d} \binom{n}{i}$ of these monomials, and we define a variable y_j for each one of them. If a cryptanalyst has access to at least $N \geq M$ keystream bits z_t, then he/she can solve the linear system of N equations for the

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14. ABSTRACT
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values of the variables y_j, and thus recover the values of $a_0, a_1, \ldots, a_{n-1}$. If d is not large, then the cryptanalyst may well be able to acquire enough keystream bits so that the system of linear equations is highly overdefined (that is, N is much larger than M).

If we use Gaussian reduction to solve the linear system, then the amount of computation required is $O\left(\binom{n}{d}^\omega\right)$, where ω is the well–known "exponent of Gaussian reduction" ($\omega = 3$ (Gauss-Jordan [11]); $\omega = \log_2 7 = 2.807$ (Strassen [12]); $\omega = 2.376$ (Coppersmith-Winograd [3])). For $n \geq 128$ and $d \sim n$, we are near the upper limits of this attack for actual systems, since the complexity grows with d.

Courtois and Meier [5] showed that if one can find a function g with small degree d_g such that $fg = 0$ or $(1 \oplus f)g = 0$, then the number of unknowns for an algebraic attack can be reduced from $\binom{n}{d_f}$ to $\binom{n}{d_g}$. That is easy to see, since $f(L^i(\mathbf{a})) = z_i$ becomes $g(L^i(\mathbf{a})) \cdot f(L^i(\mathbf{a})) = 0 = z_i\, g(L^i(\mathbf{a}))$, and so, we get the equations $g(L^i(\mathbf{a})) = 0$, whenever the intercepted $z_i \neq 0$. That gives us a reduction in complexity, from $O\left(\binom{n}{d_f}^\omega\right)$ to $O\left(\binom{n}{d_g}^\omega\right)$. Therefore, it is necessary to have a fast computation of a low(est) degree annihilator of the combiner f.

2 Background and Notation

2.1 Introduction

The objects of our study are Boolean functions $f : \mathbb{V}_n \to \mathbb{F}_2$ (see [6] for more properties on Boolean functions).

Definition 2.1. The **degree** d of a **term** $x_{i_1} x_{i_2} \ldots x_{i_d}$ is the number of distinct variables in that term, where $i_j \in \{1, 2, \ldots, n\}$.

Definition 2.2. The algebraic normal form or **ANF** of function $f(x_1, x_2, \ldots, x_n)$ is a polynomial expression of f consisting of the exclusive OR of terms.

The ANF of a function is often referred to as the *positive polarity Reed-Muller* form.

Definition 2.3. The **degree**, $\deg(f)$, of function $f(x_1, x_2, \ldots, x_n)$ is the largest degree among all the terms in the ANF of f.

Example 2.1. The ANF of the majority function (in 3 variables) is $f(x_1, x_2, x_3) = x_1 x_2 \oplus x_1 x_3 \oplus x_2 x_3$. Its degree is 2.

Definition 2.4. Function a is an **annihilator** of function f if and only if $a \cdot f = 0$, where $a \neq 0$.

Note that $\bar{f}$ is an annihilator of f. Further, if a is annihilator of f, so also is α, where $\alpha \leq a$ ('$\leq$' is a partial order on the set of vectors of the same dimension, that is, $(\alpha_i)_i \leq (\beta_i)_i$ if and only if $\alpha_i \leq \beta_i, \forall i$).

Definition 2.5. Function f has **algebraic immunity** $k = \min\{\deg(a) | a \text{ is an annihilator of } f \text{ or } \bar{f}\}$.

Example 2.2. The annihilators of $f(x_1, x_2, x_3) = x_1 x_2 \oplus x_1 x_3 \oplus x_2 x_3$ include $\bar{f}$ and all g such that $g \leq \bar{f}$, excluding the constant 0 function. In all, there are 15 annihilators of f and 15 annihilators of $\bar{f}$. Among these 30 functions, the minimum degree is 2. Thus, $f(x_1, x_2, x_3) = x_1 x_2 \oplus x_1 x_3 \oplus x_2 x_3$ has algebraic immunity 2. Table 1 shows all 15 annihilators of the 3-variable majority function.

Example 2.3. Let $n = 4$. The function $f = x_1 x_2 x_3 x_4$ has the highest degree, that is, 4. Function $a = \bar{x}_1 = x_1 \oplus 1$ annihilates f, since $a \cdot f = \bar{x}_1 x_1 x_2 x_3 x_4 = 0$. Since, $\bar{x}_1$ has degree 1 and there exists no annihilator of f of degree 0, the algebraic immunity of f is 1. To reach this conclusion, it is not necessary to check the annihilators of $\bar{f}$, since the only annihilator of $\bar{f}$ is f, which has degree 4.

We can immediately state

Lemma 2.1. The algebraic immunity of a function $f(x_1, x_2, \ldots, x_n)$ is identical to the algebraic immunity of $\bar{f}$.

Table 1: Functions that annihilate the 3-variable majority function and their degree

$x_1x_2x_3$	f	α_1	α_2	α_3	α_4	α_5	α_6	α_7	α_8	α_9	α_{10}	α_{11}	α_{12}	α_{13}	α_{14}	α_{15}
000	0	1	0	1	0	1	0	1	0	1	0	1	0	1	0	1
001	0	1	1	0	0	1	1	0	0	1	1	0	0	1	1	0
010	0	1	1	1	1	0	0	0	0	1	1	1	1	0	0	0
011	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
100	0	1	1	1	1	1	1	1	1	0	0	0	0	0	0	0
101	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
110	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
111	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
Degree		2	3	3	2	3	2	2	3	3	2	2	3	2	3	3

More specifically,

Lemma 2.2. *The algebraic immunity of a function $f(x_1, x_2, \ldots, x_n)$ is $\min\{\deg(\alpha)|\alpha \leq \bar{f} \text{ or } \alpha \leq f\}$.*

Proof The hypothesis follows immediately from the observation that $\{\alpha|\alpha \leq \bar{f} \text{ or } \alpha \leq f\}$ is the set of all annihilators of f. ∎

Lemma 2.2 is similar to Definition 2.5. However, there is an important difference. Lemma 2.2 admits an algorithm for determining the algebraic immunity of a function f. Specifically, examine the degree of each function α such that $\alpha \leq \bar{f}$ and determine the minimum degree of the ANF among all α. This requires the examination of $2^{2^n - wt(f)} - 1$ functions, where $wt(f)$ is the number of 1's in the truth table of f, since $\bar{f}$ has $2^n - wt(f)$ 1's in its truth table. In forming an annihilator, each 1 can be retained or set to 0. The '−1' accounts for the case where all 1's are set to 0, which is not an annihilator. The following result is essential to the *efficient* computation of algebraic immunity.

Lemma 2.3. *[5, 9] The algebraic immunity $AI(f)$ of a function $f(x_1, x_2, \ldots, x_n)$ is bounded; $AI(f) \leq \lceil \frac{n}{2} \rceil$.*

3 Computation of Algebraic Immunity

3.1 Row Echelon Reduction Method for Algebraic Immunity Computation

There are several methods for computing the algebraic immunity of a Boolean function f. Besides the brute force algorithm (check every function to see if it is an annihilator of the given f, or its complement), one can also use an approach in which one can identify directly the annihilators with high degree. Since our attempt is to implement the algebraic immunity computation on a reconfigurable computer, we have not implemented the more recent algorithm of Armchnecht et al. [1], which also deals with *fast* algebraic attack issues. Our approach is based on a simpler version of that linear algebra approach. This enabled us to implement it on the SRC-6 reconfigurable computer and to display the AI profiles for all functions on $\mathbb{F}_2^n$, for $2 \leq n \leq 5$. A similar approach has been used to compute algebraic immunity on a conventional processor [5, 9] . Our implementation is the first known using Verilog on an FPGA.

Here, we create the ANF of a minterm corresponding to each 1 in the truth table of the function. Our approach to solving this system is to express this in reduced row echelon form using Gaussian elimination that is based on two elementary row operations: 1. interchange two rows, and 2. add one row to another row. A simple test applied to the reduced row echelon form determines if there is an annihilator of some specified degree.

3.2 Example

To illustrate, consider solving the algebraic immunity of the majority function $f(x_1, x_2, x_3) = x_1x_2 \oplus x_1x_3 \oplus x_2x_3$. The top half of Table 2 shows the minterm canonical form of $\bar{f}$. Here, the

first (leftmost) column represents all binary three tuples on three variables. The second column contains the truth table of the complement function $\bar{f}$, which is expressed as $\bar{x}_1\bar{x}_2\bar{x}_3 \vee \bar{x}_1\bar{x}_2 x_3 \vee \bar{x}_1 x_2 \bar{x}_3 \vee x_1\bar{x}_2\bar{x}_3$, or more compactly, as the sum of minterms $m_0 \vee m_1 \vee m_2 \vee m_4$. This represents the annihilator of f with the most 1's.

The columns are labeled by all possible terms in the ANF of an annihilator. Then, 1's are inserted into the table to represent the ANF of the minterms. For example, since the top minterm, $m_0 = \bar{x}_1\bar{x}_2\bar{x}_3 = (x_1\oplus 1)(x_2\oplus 1)(x_3\oplus 1) = x_1 x_2 x_3 \oplus x_2 x_3 \oplus x_1 x_3 \oplus x_1 x_2 \oplus x_3 \oplus x_2 \oplus x_1 \oplus 1$, its ANF has all possible terms, and so, there is a 1 in every column of this row.

Note that we can obtain the ANF of some combination of minterms as the exclusive OR of various rows in the top half of Table 2. This follows from the observation that $m_i \vee m_j = m_i \oplus m_j$. For example, one annihilator a is $a = \bar{x}_1 x_2 \bar{x}_3 \vee \bar{x}_1\bar{x}_2\bar{x}_3$, and so the ANF of a is generated by simply exclusive ORing the rows associated with these two minterms.

Table 2: Functions that annihilate the 3-variable majority function

ANF Coefficient →			c_7	c_6	c_5	c_4	c_3	c_2	c_1	c_0	Minterms
Index	$x_1 x_2 x_3$	$\bar{f}$	$x_1 x_2 x_3$	$x_2 x_3$	$x_1 x_3$	$x_1 x_2$	x_3	x_2	x_1	1	
Original											
0	000	1	1	1	1	1	1	1	1	1	m_0
1	001	1	1	1	1	0	1	0	0	0	m_1
2	010	1	1	1	0	1	0	1	0	0	m_2
3	011	0	0	0	0	0	0	0	0	0	
4	100	1	1	0	1	1	0	0	1	0	m_4
5	101	0	0	0	0	0	0	0	0	0	
6	110	0	0	0	0	0	0	0	0	0	
7	111	0	0	0	0	0	0	0	0	0	
Reduced Row Echelon Form											
0	-	-	1	0	0	0	1	1	1	0	$m_1\oplus m_2\oplus m_4$
1	-	-	0	1	0	0	1	1	0	1	$m_0 \oplus m_4$
2	-	-	0	0	1	0	1	0	1	1	$m_0 \oplus m_2$
3	-	-	0	0	0	1	0	1	1	1	$m_0 \oplus m_1$
4	-	-	0	0	0	0	0	0	0	0	
5	-	-	0	0	0	0	0	0	0	0	
6	-	-	0	0	0	0	0	0	0	0	
7	-	-	0	0	0	0	0	0	0	0	

3.3 Elementary Row Operations

Consider a 0-1 matrix and two row operations: 1. interchange one row with another, and 2. replace one row by the exclusive OR of that row with any other row. Using elementary row operations, we seek to create columns, starting with the left column with *only one* 1 (called a *pivot*).

3.4 Row Echelon Form

Definition 3.6. *A 0-1 matrix is in **row echelon form** iff all nonzero rows (if they exist) are above any rows of all zeroes, and the leading coefficient (pivot) of a nonzero row is always strictly to the right of the leading coefficient of the row above it.*

Definition 3.7. *[14] A 0-1 matrix is in **reduced row echelon form** iff it is in row echelon form and each leading 1 (pivot) is the only 1 in its column.*

Consider Table 2. The top half shows the truth table of $\bar{f}$. For each 1 (minterm - m_0, m_1, m_2, and m_4) in the $\bar{f}$ column (third column), the ANF of that minterm is expressed across the rows. To form an annihilator of f, we must combine one or more minterms using the exclusive OR operation. The bottom half of Table 2 shows the reduced row echelon form of the top half. The row operations we used to derive the bottom half from the top half can be inferred from the rightmost column. For instance, the entry $m_0 \oplus m_1$ in the bottom half of the table indicates

that the rows labeled m_0 and m_1 in the top half of the table were combined using the exclusive OR operation.

Note that, like the top half of Table 2, the rows in the reduced row echelon form combine to form *any* annihilator of the original function. This follows from the fact that any single minterm can be formed as the exclusive OR of rows in the reduced row echelon form. For example, m_1 is obtained as the exclusive OR of the top three rows of the reduced row echelon form.

The advantage of the reduced row echelon form is that we can simply inspect the rows to determine the annihilators of lowest degree. For example, in the reduced row echelon form, the top row represents an annihilator of degree 3, since there is a 1 in the column associated with $x_1x_2x_3$. Since the pivot point has the *only* 1 in this row, the only way to form an annihilator of degree 3 is to include this row.

The other three rows each have a pivot in a column associated with a degree 2 term. And, the only way to have a degree 2 term is to involve at least one of these rows. Since there are no other rows with a pivot point in a degree 1 or 0 term, we can conclude that there exist *no* annihilators of degree 1 or 0. Thus, the lowest degree of an annihilator of $f\ (= x_1x_2 \oplus x_1x_3 \oplus x_2x_3)$ is 2.

3.5 Steps to Reduce the Computation Time

The matrices for which we seek a reduced row echelon form can be large. For example, each matrix has 2^n rows, of which we manipulate only those with 1's in the function. Potentially, there are also 2^n columns. However, we can reduce the columns we need to examine by a few observations. Recall that no function has an AI greater than $\lceil \frac{n}{2} \rceil$. Thus, we need consider only those columns corresponding to ANF terms where there are $\lceil \frac{n}{2} \rceil$ or fewer variables. However, it is not necessary to consider columns corresponding to terms with exactly $\lceil \frac{n}{2} \rceil$ variables. This is because if no annihilators are found for a function f (or its complement) of degree $\lceil \frac{n}{2} \rceil - 1$ or less, it must have an AI of $\lceil \frac{n}{2} \rceil$.

We can reduce the computation of the AI of a function by another observation. If a degree 1 annihilator for a function is found, there is no need to analyze its complement. Even if the complement has no annihilators of degree 1, the function itself has AI of 1. On the other hand, finding an annihilator of degree $\lceil \frac{n}{2} \rceil$ requires the analysis of its complement for annihilators of smaller degree.

4 Results

4.1 Approach

A Verilog program was written to implement the row echelon conversion process described above. It runs on an SRC-6 reconfigurable computer from SRC Computers, Inc. and uses the Xilinx Virtex2p (Virtex2 Pro) XC2VP100 FPGA with Package FF1696 and Speed Grade -5. Table 3 compares the average time in computing the AI of an n-variable function on this FPGA with the rate of a typical microprocessor. In this case, we chose the Intel®Core$^{\text{TM}}$2 Duo P8400 processor running at 2.26 GHz. This processor runs Windows 7 and has 4 GB of RAM. The code was compiled using Code::Blocks 10.05. The data shown is from a C program that also implements the row echelon conversion process.

4.2 Computation Times

Table 3 compares the computation times for AI when done on the SRC-6 reconfigurable computer and on an Intel® Core$^{\text{TM}}$2 Duo P8400 processor. The second, third, and fourth columns show the performance of the SRC-6 and the next two columns show the performance on the Intel® Core$^{\text{TM}}$2 Duo P8400 processor. The last column shows the speedup of the SRC-6 over the Intel® Core$^{\text{TM}}$2 Duo P8400 processor. The second column shows the average number of 100 MHz clocks needed by the SRC-6. The third column shows the average number of functions per second. The fourth column shows the number of functions. In the case of $n \le 5$, all functions were enumerated, and in the case of $n = 6$, a subset of random functions was enumerated. The fifth column shows the average number of functions per second on the Intel® Core$^{\text{TM}}$2 Duo P8400 processor, while the sixth column shows the number of functions. The last column shows the speedup of the reconfigurable computer over the Intel® Core$^{\text{TM}}$2 Duo P8400 processor.

For example, for $n = 5$, the SRC-6 reconfigurable computer is 4.9 times faster than the Intel® processor. For $n = 4$, the SRC-6 is 1.9 times faster. However, for $n = 6$, the processor is actually faster than the reconfigurable computer. In the case of $n = 6$, a sample size of 25,000,000 was used for the SRC-6 and 500,000,000 for the Intel® Core™2 Duo P8400 processor. For all lower values of n, exhaustive enumeration was performed.

Table 3: Comparison of the computation times for enumerating the AI of n-variable functions on the SRC-6 reconfigurable computer versus an Intel® Core™2 Duo P8400 microprocessor

	SRC-6 Reconfigurable Comp.			Intel®Processor		
n	Clocks per function	Functions per second	# samples	Functions per second	# samples	Speed-up
2	46.3	2,162,162	16,000,000*	4,186,290	16,000,000*	0.5
3	70.7	1,414,130	25,600,000*	1,317,076	25,600,000*	1.1
4	75.5	880,558	65,536,000*	458,274	65,536,000*	1.9
5	348.4	287,012	4,294,967,296*	59,029	4,294,967,296*	4.9
6	78.0	12,823	25,000,000	17,699	10,000,000,000	0.7

* Exhaustive enumeration - All n-variable functions were enumerated.

4.3 Comparing the Row Echelon Method to Brute Force

Table 4 compares the row echelon method, which involves the solution of simultaneous equations with the brute force method discussed earlier for the case of $n = 4$. In both cases, 65,536 functions were considered, all 4-variable functions. The last row shows that the row echelon method is able to process 1,325,000 functions per second verses 81,160 for the brute force method, resulting in 16.3 times the throughput

Table 4: Comparing the Brute Force Method With the Row Echelon Method on 4-Variable Functions

	Brute Force	Row Echelon
# Functions	65,536	65,536
Total Time (sec.)	0.807	0.050
Total Clocks	80,748,733	4,946,111
Clocks Per Function	1,232.1	75.5
Functions Per Second	81,160	1,325,000

4.4 Distribution of Algebraic Immunity to Functions

Table 5 shows the number of functions with various algebraic immunities for $2 \leq n \leq 6$. This extends the results of [13] to $n = 5$. In our case, the use of a reconfigurable computer allows this extension. The entries shown in **bold** in the column for $AI = 5$ are exact values for previously unknown values. The entries shown in ***bold and italics*** for $AI = 6$ are approximate values for previously unknown values. In this case, the approximate values were determined by a Monte Carlo method in which 500,000,000 random 6-variable functions were generated (or $2.7 \times 10^{-9}\%$ of the total number of functions) and their algebraic immunity computed. For $n = 5$ and $n = 6$, the number of functions with algebraic immunity 1 are known. However, Table 5 shows the value 0 for the number of functions with algebraic immunity 0 (there are actually 2, the exclusive OR function and its complement). This is because the Monte Carlo method produced no functions with an AI of 0. The *italicized* value, *1,114,183,342,052*, in Table 5 is an estimate of the number of 6-variable functions with algebraic immunity 1. To show the accuracy of the Monte Carlo method, compare this to the previously known exact value 1,081,682,871,734 [13]. The estimated value is 3% greater than the exact value.

Table 5: The number of n-variable functions distributed according to algebraic immunity for $2 \leq n \leq 6$.

AI\n	2	3	4	5	6
0	2	2	2	2	(2) 0
1	14	198	10,582	7,666,550	(1,081,682,871,734) *1,114,183,342,052*
2	0	56	54,952	**4,089,535,624**	***1,269,840,659,739,507,264***
3	0	0	0	**197,765,120**	***17,176,902,299,786,702,300***
TOTAL	16	256	65,536	4,294,967,296	18,446,744,073,709,551,616

Bold entries are previously unknown. ***Bold and italicized*** entries are estimates to previously unknown values.

4.5   Resources Used

Table 6 shows the frequency achieved on the SRC-6 and the number of LUTs and flip-flops needed in the realization of the AI computation for various n. The frequency ranges from 109.4 MHz at $n = 4$ to 87.5 MHz for $n = 6$. Since the SRC-6 runs at 100 MHz, the 87.5 MHz value is cause for concern. However, the system works well at this frequency. For all values of n, the number of LUTs, slice flip-flops, and occupied slices were well within FPGA limits. Indeed, among all three parameters and all values of n, the highest percentage was 11%.

Table 6: Frequency and resources used to realize the AI computation on the SRC-6's Xilinx Virtex2p (Virtex2 Pro) XC2VP100 FPGA.

n	Freq. (MHz)	# of LUTs	Total # of Slice FFs	# of occu- pied Slices
2	103.1	2,066(2%)	2,977(3%)	2,089(4%)
3	113.0	2,199(2%)	3,011(3%)	2,157(4%)
4	109.4	2,343(2%)	2,760(3%)	2,120(4%)
5	100.9	5,037(5%)	4,110(4%)	3,780(8%)
6	87.5	8,990(10%)	3,235(3%)	5,060(11%)

5   Concluding Remarks

We show that a reconfigurable computer can be programmed to efficiently compute the algebraic immunity of a logic function. Specifically, we show a 4.9 times speedup over the computation time of a conventional processor. This is encouraging given that algebraic immunity is one of the most complex cryptographic properties to compute. This is the third cryptographic property that has benefited from the highly efficient, parallel nature of the reconfigurable computer. The interested reader may wish to consult two previous papers: nonlinearity [10] and correlation immunity [7].

6   Acknowledgments

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